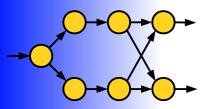


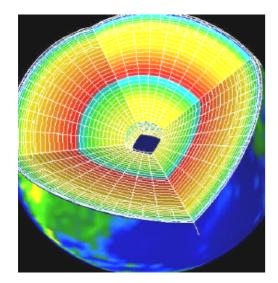
On Runtime Systems for Task-based Programming on Heterogeneous Platforms

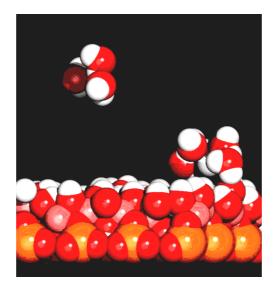
Samuel Thibault SATANAS/STORM

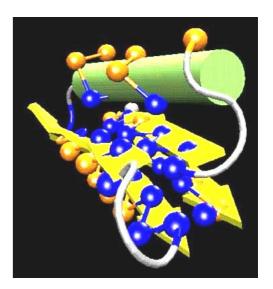


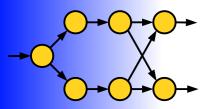


Simulation complements theory and experimentation

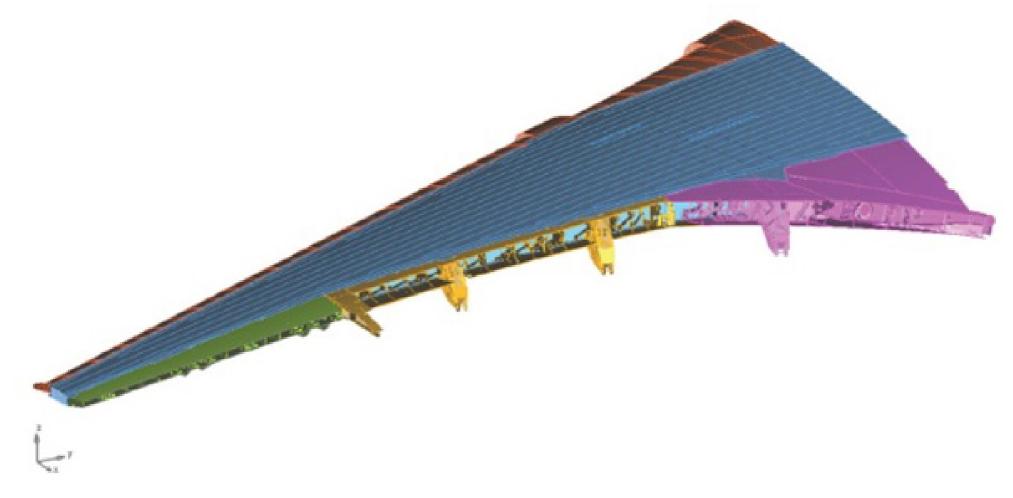


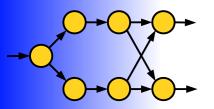






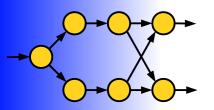
Simulation complements theory and experimentation, e.g. A350





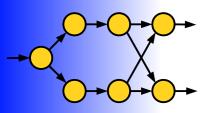
Simulation complements theory and experimentation, e.g. A350





Simulation complements theory and experimentation But requires huge computational power



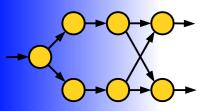


Summit supercomputer

- #1 on LINPACK benchmark
- 143PF
- 4608 nodes
 - 2 IBM Power9 CPUs
 - 6 NVIDIA V100 GPUs
 - 512 GB DDR + 96 GB HBM2 mem
- IB 100G network
- 13MW

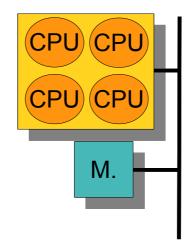


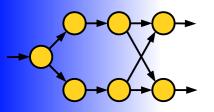
From ORNL



Classical parallel programming

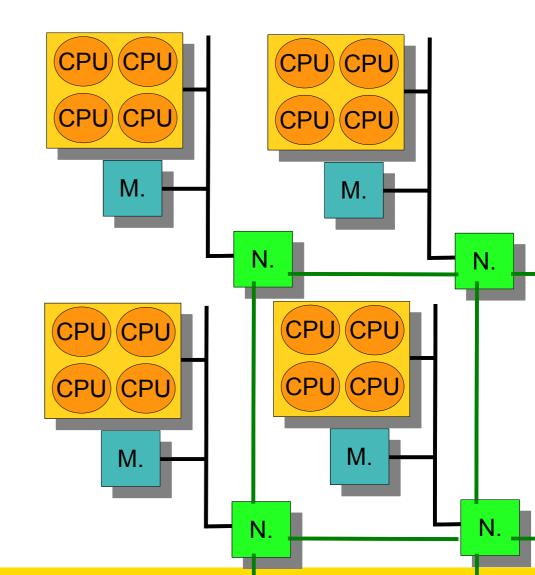
• threads

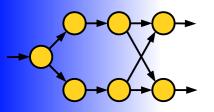




Classical parallel programming

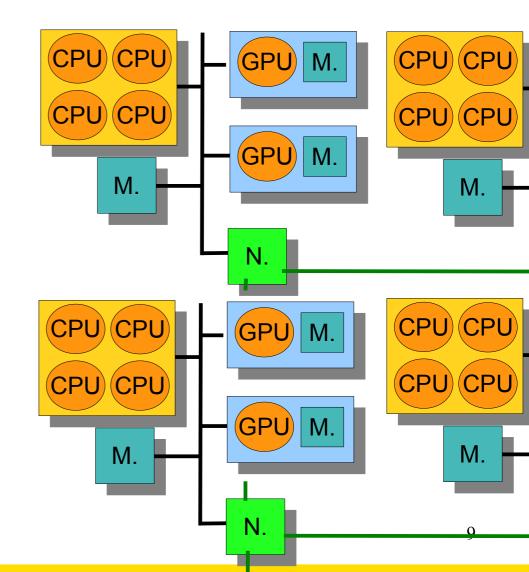
- threads
- MPI+threads

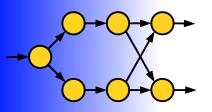




Classical parallel programming

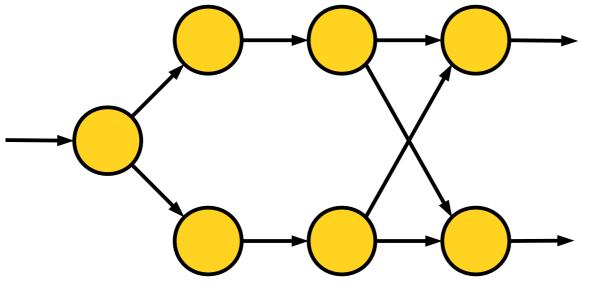
- threads
- MPI+threads
- MPI+threads+CUDA
- Managing interaction between the three?



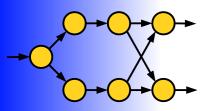


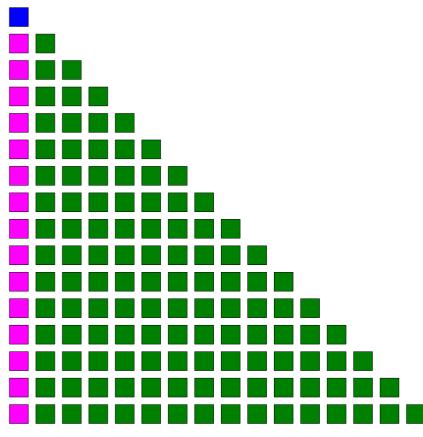
Task graphs

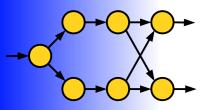
• Well-studied for scheduling parallelism (since '60s!)



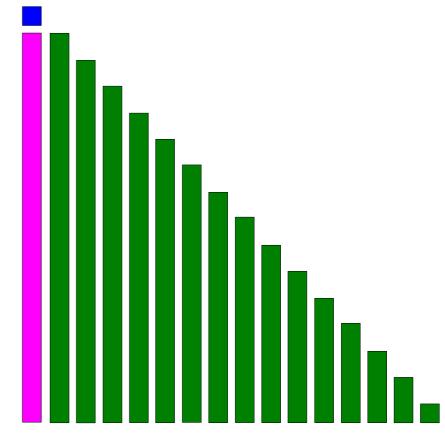
- Runtimes existing for a long time
 - e.g. Cilk, Athapascan-1 (end '90s)
- Until recently, not really actually used for scientific computation
 - Heterogeneous architectures pushed for them

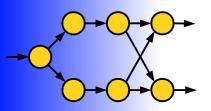




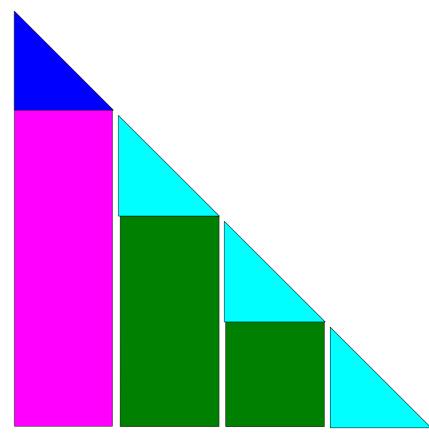


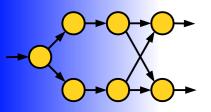
LINPACK: vector computers (end '70s, '80s)



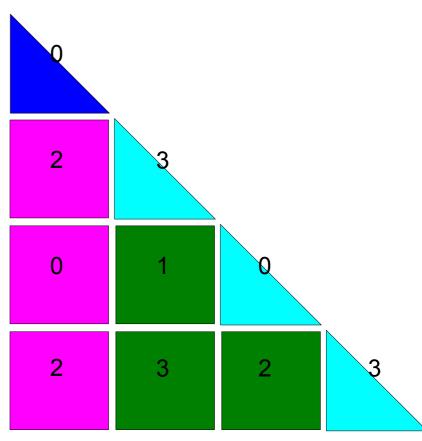


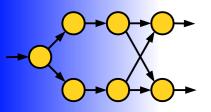
- LINPACK: vector computers (end '70s, '80s)
- LAPACK: cache-aware ('90s), blocked operation



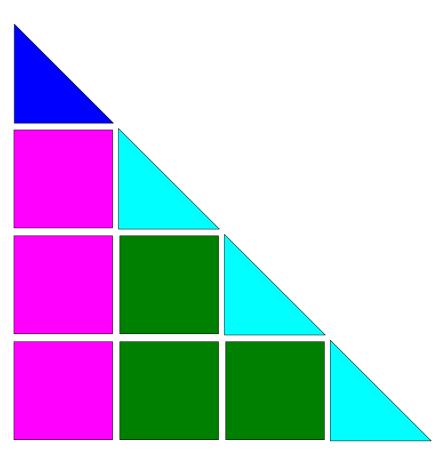


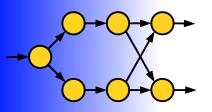
- LINPACK: vector computers (end '70s, '80s)
- LAPACK: cache-aware ('90s), blocked operation
- ScaLAPACK: distributed ('90s)
 - 2D block-cyclic distribution





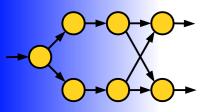
- LINPACK: vector computers (end '70s, '80s)
- LAPACK: cache-aware ('90s), blocked operation
- ScaLAPACK: distributed ('90s)
 - 2D block-cyclic distribution
- PLASMA: task graph (~'08)
 - + StarPU runtime + MAGMA GPU kernels





What happened?

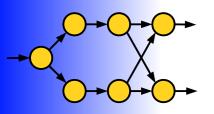
- Application writers wanted to keep control
 - Know their hardware
 - Strict ordering
 - Bulk Synchronous Parallelism
 - Hand-tuned pipelining
- Is that maintainable?
 - New hardware, rewrite?
- GPUs pushed to task graphs
 - MPI+threads+CUDA...



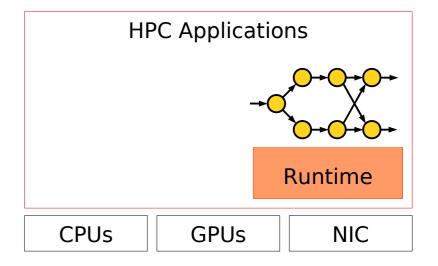
• From static control...

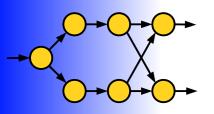
HPC Applications

CPUs	GPUs	NIC

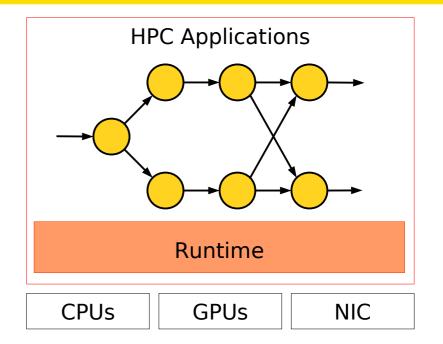


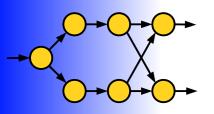
- From static control...
- To some dynamic control...



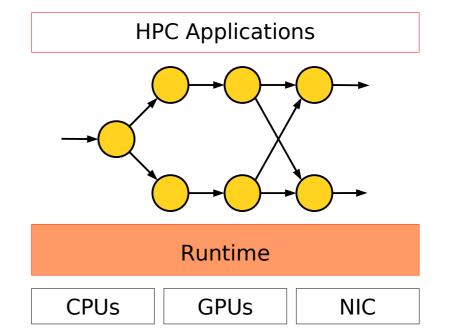


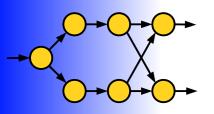
- From static control...
- To some dynamic control...
- To giving it all to the runtime...



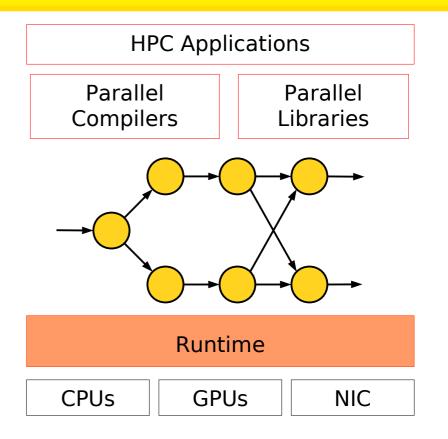


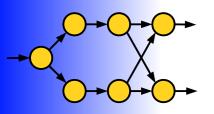
- From static control...
- To some dynamic control...
- To giving it all to the runtime...
- Or even an external runtime



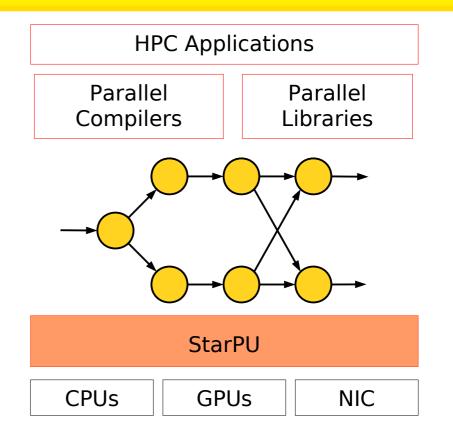


- From static control...
- To some dynamic control...
- To giving it all to the runtime...
- Or even an external runtime
- Possibly through intermediate layers
- Many projects leaned to this

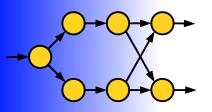




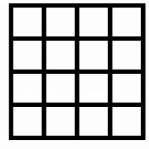
- From static control...
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- To giving it all to the runtime...
- Or even an external runtime
- Possibly through intermediate layers
- Many projects leaned to this

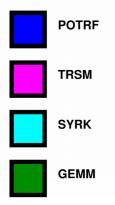


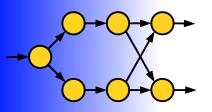
- StarPU from the start (2008) external
 - [AugonnetPhD11]



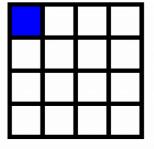
Sequential Task Flow (STF)

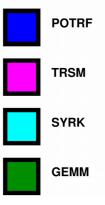


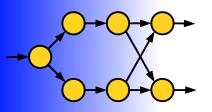




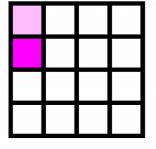
Sequential Task Flow (STF)

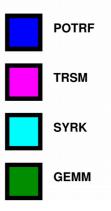


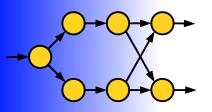




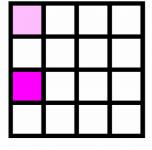
Sequential Task Flow (STF)



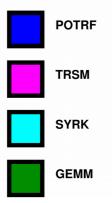


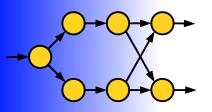


Sequential Task Flow (STF)

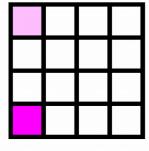


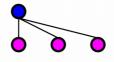


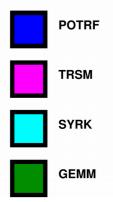


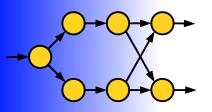


Sequential Task Flow (STF)

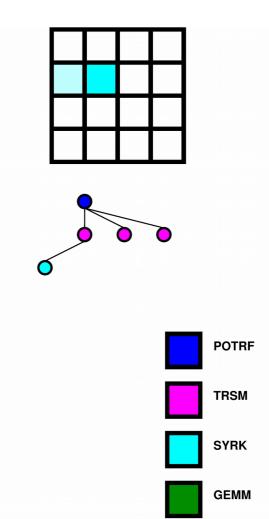


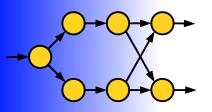




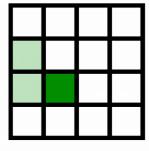


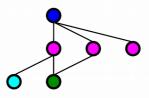
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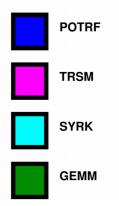


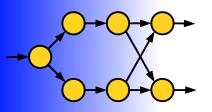


Sequential Task Flow (STF)

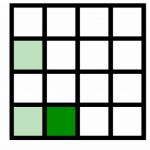


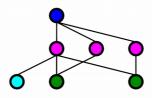


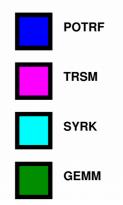


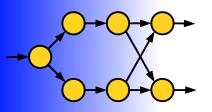


Sequential Task Flow (STF)

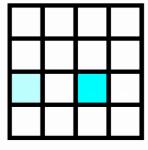


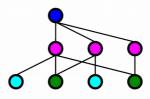


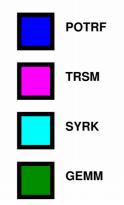


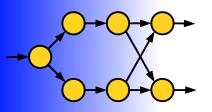


Sequential Task Flow (STF)

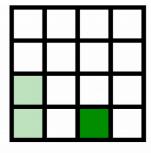


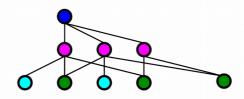


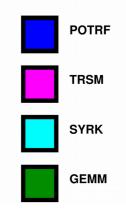


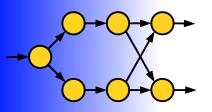


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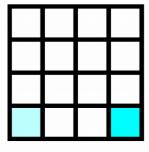


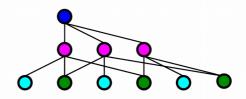


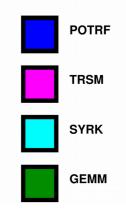


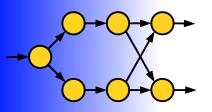


Sequential Task Flow (STF)

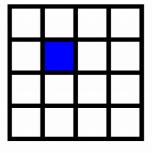


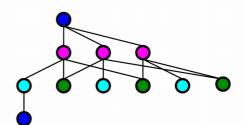


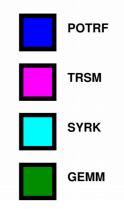


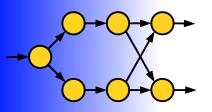


Sequential Task Flow (STF)





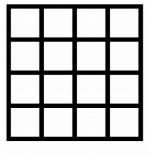


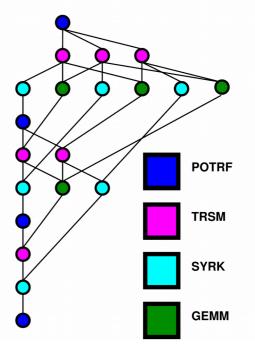


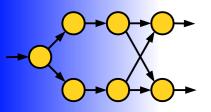
Sequential Task Flow (STF)

Cholesky example from PLASMA

task_wait_for_all();

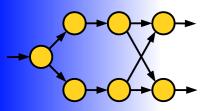






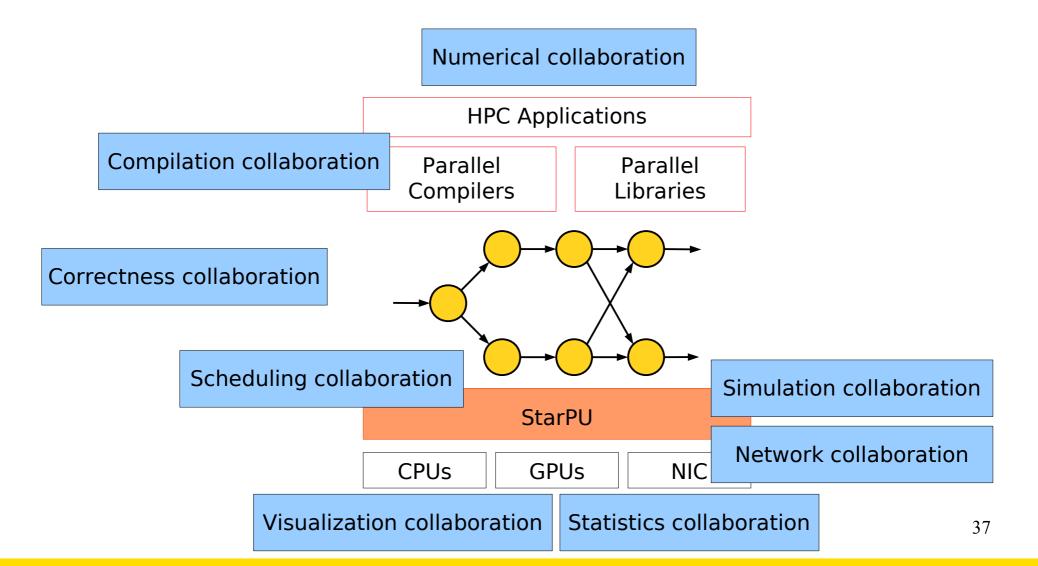
Sequential Task Flow (STF)

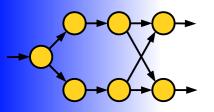
- Sequential-looking source code
 - Just expresses the algorithm
- Can debug sequential version
- Runtime will handle parallel execution



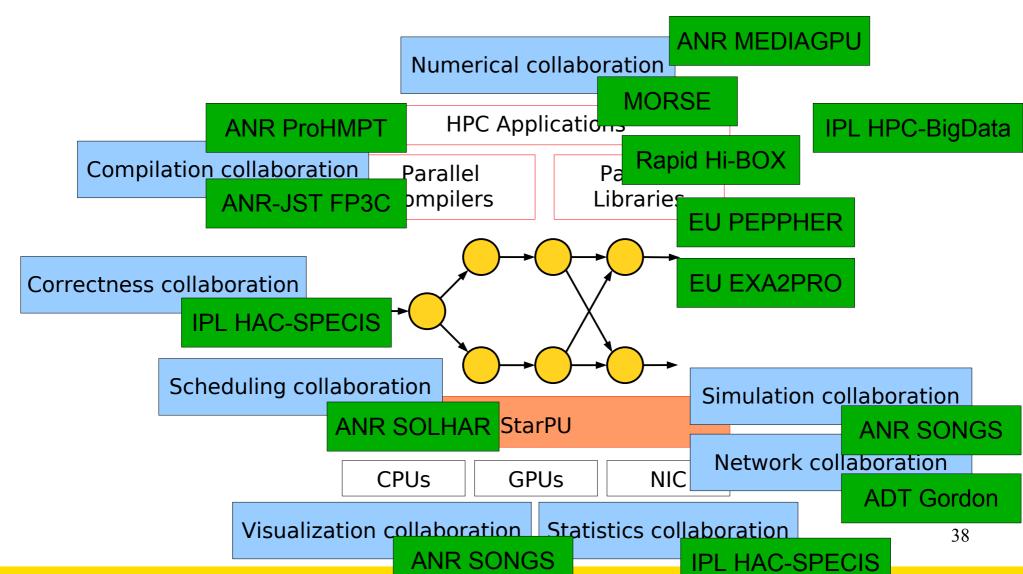
Task graphs as central notion

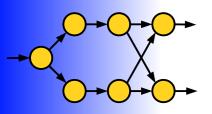
Task graphs as a bridge between various expertise





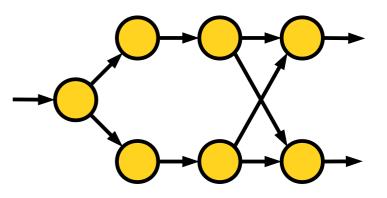
Task graphs as a bridge between various expertise

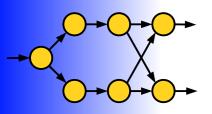




Today's agenda

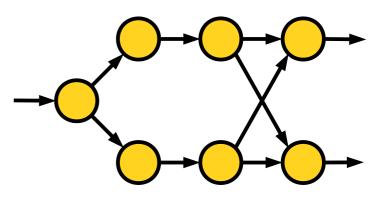
- Task graph scheduling
 - What is a runtime scheduler?
 - Beyond classical HEFT
 - Involving theoreticians
- Dividable tasks
- Distributed execution

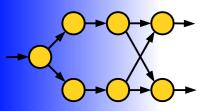




Today's agenda

- Task graph scheduling
 - What is a runtime scheduler?
 - Beyond classical HEFT
 - Involving theoreticians
- Dividable tasks
- Distributed execution

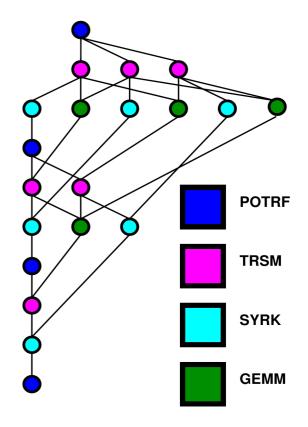


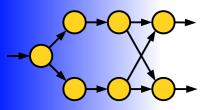


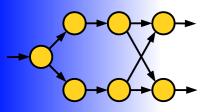
Task graph scheduling

Target: completion time, usually (could also be energy)

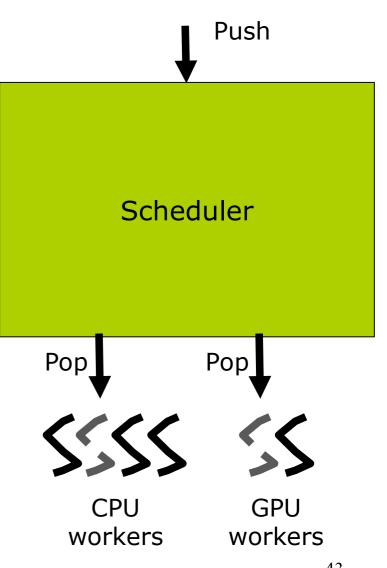
- Care of critical path
 - Task priorities
- Care of leveraging accelerators
 - Task duration
- Care of data transfers
 - Transfer penalty

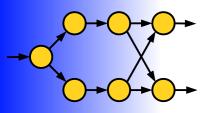




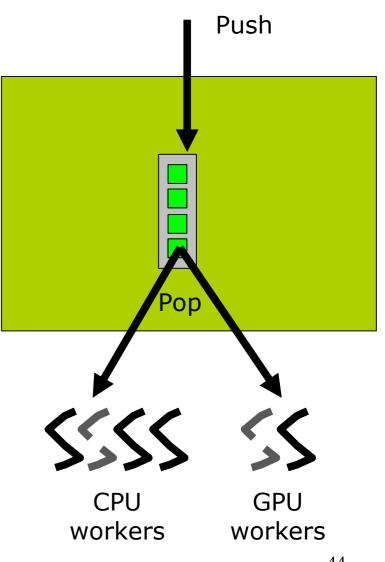


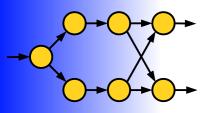
• Ready tasks



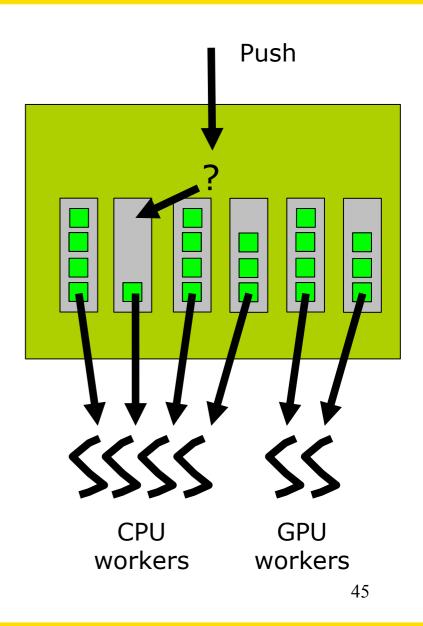


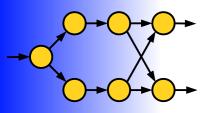
- Ready tasks
- Single eager list



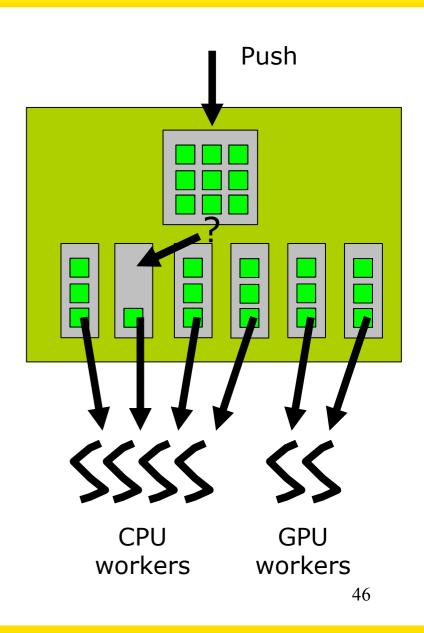


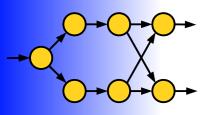
- Ready tasks
- Single eager list
- Early decision





- Ready tasks
- Single eager list
- Early decision
- But not too early

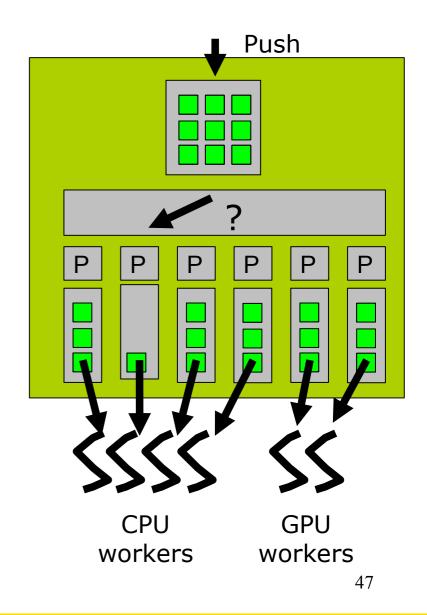


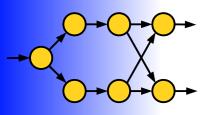


- Ready tasks
- Single eager list
- Early decision
- But not too early

Move to component-based

- [Archipoff13, SergentPhD16]
- Tasks pushed/pulled

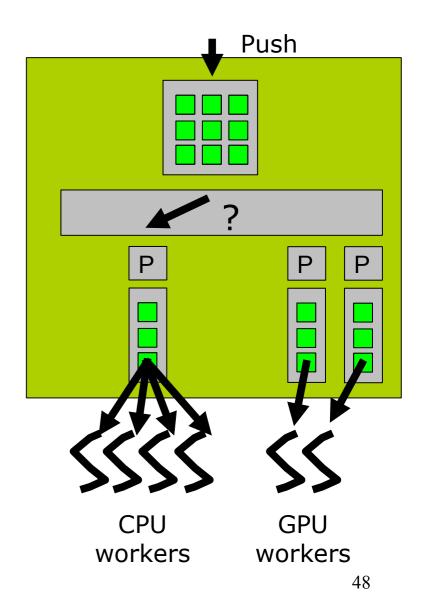


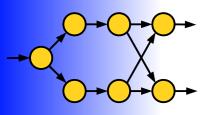


- Ready tasks
- Single eager list
- Early decision
- But not too early

Move to component-based

- [Archipoff13, SergentPhD16]
- Tasks pushed/pulled

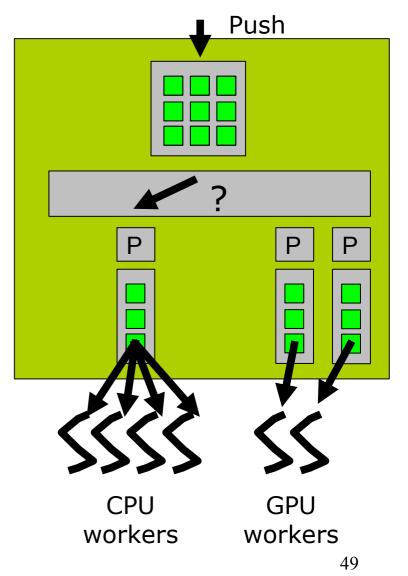


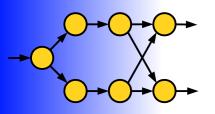


- Ready tasks
- Single eager list
- Early decision
- But not too early

Move to component-based

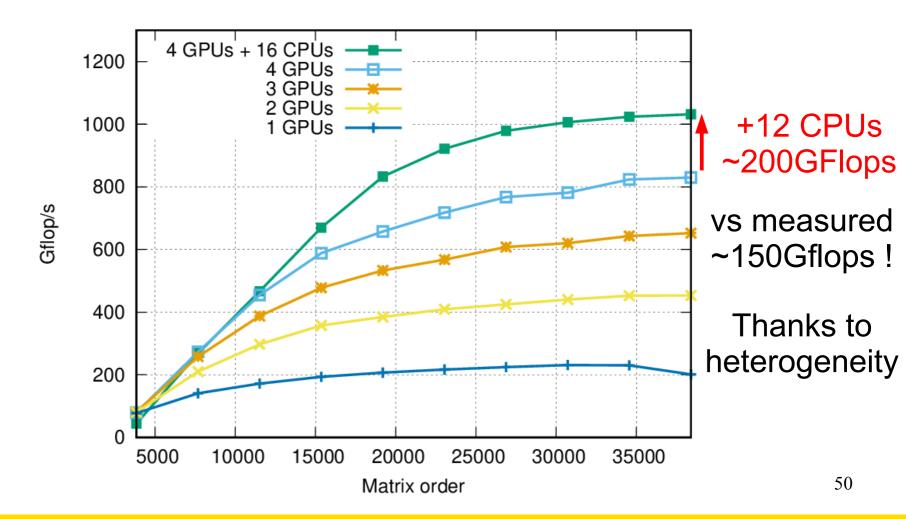
- [Archipoff13, SergentPhD16]
- Tasks pushed/pulled
- Would welcome correctness model

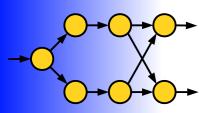




Typical obtained performance

- With classical HEFT-like scheduling heuristic (dmdas)
 - e.g. QR factorization

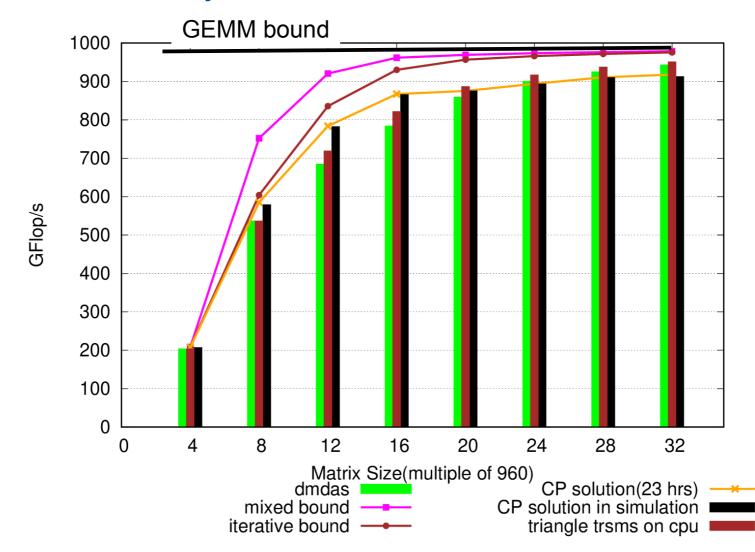


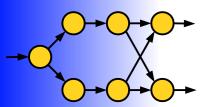


- HEFT-like heuristic dmdas is strong
 - But can do better
- [KumarPhD17], co-advised with
 - Theoreticians O. Beaumont, L. Eyraud-Dubois
 - Numerical analyst E. Agullo
 - Runtime really acted as a bridge between us
- Focused on Cholesky factorization
 - Findings applicable to dense linear algebra in general
- Improved performance bounds
- Injected static knowledge of the application into runtime scheduler

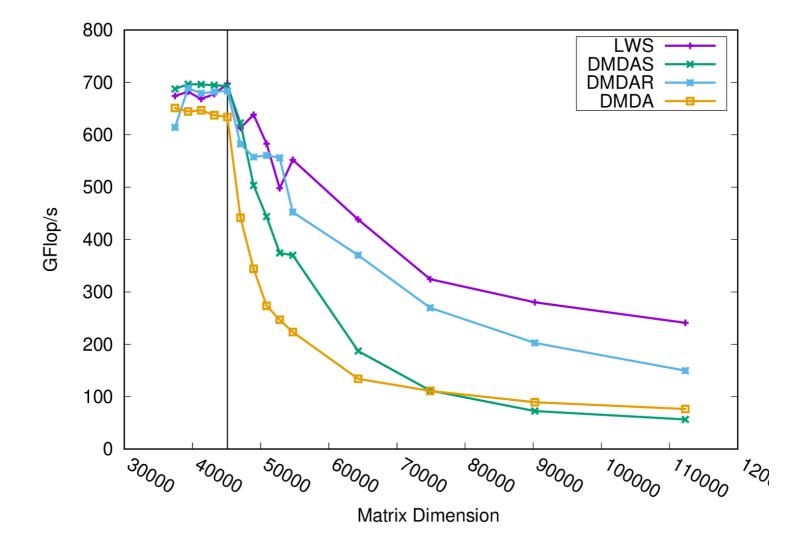
Scheduling beyond state of the art

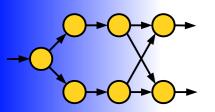
Dense Cholesky factorization





• Dense Cholesky beyond memory size (out-of-core)





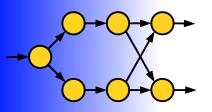
Scheduling beyond state of the art

Hi-BOX project with Airbus

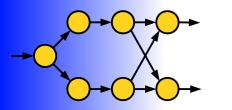
- H-matrices
 - Hierarchically compressed blocks
 - Growing data
- e.g. 1600 GB result on 256 GB system
- Spends half the time exchanging data with disk

Still need to find proper compromise between

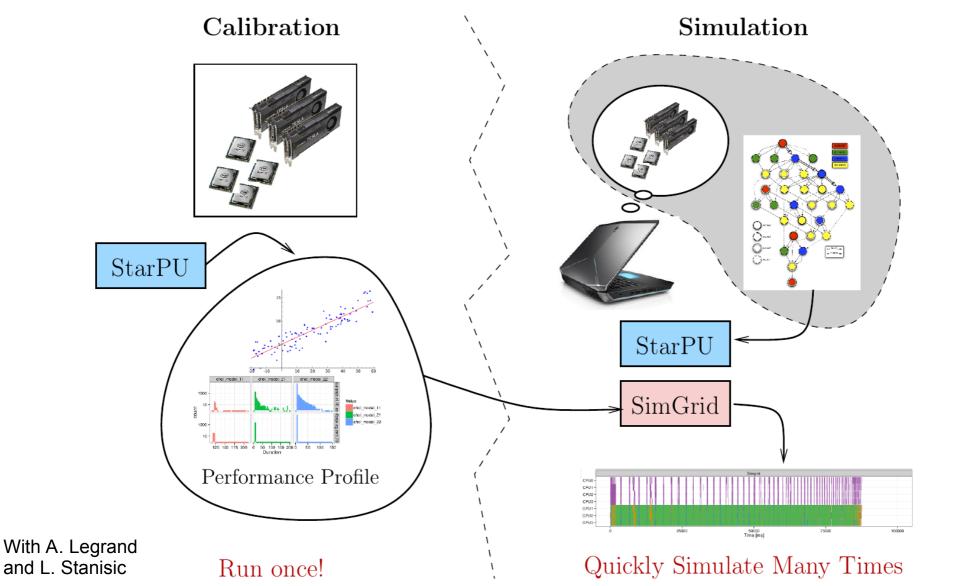
- critical path,
- acceleration,
- and data transfer

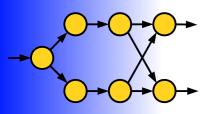


- Component-based schedulers should help
- Running actual application + runtime remains a pain
 - Getting access to target platforms
 - Installing software
 - And dependencies!
- \rightarrow Simulation

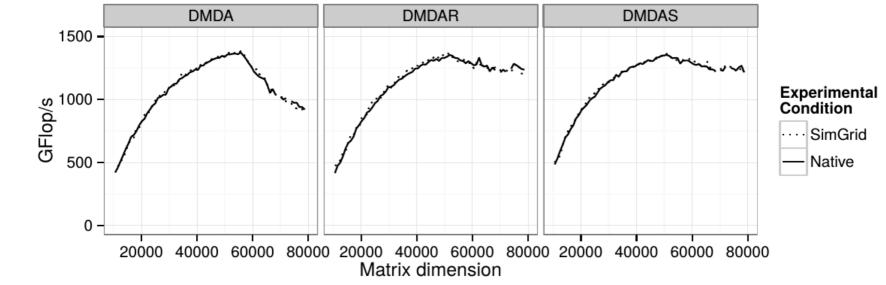


Simulation with SimGrid





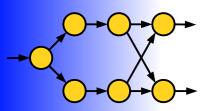
Simulation with SimGrid



Execute real application in simulation mode

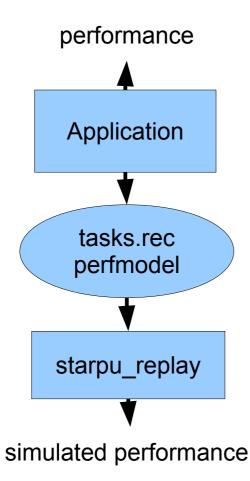
- Way faster execution time
- Reproducible experiments
- No need to run on target system
- Can change system architecture

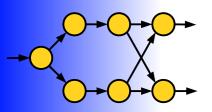
But still requires to build real application



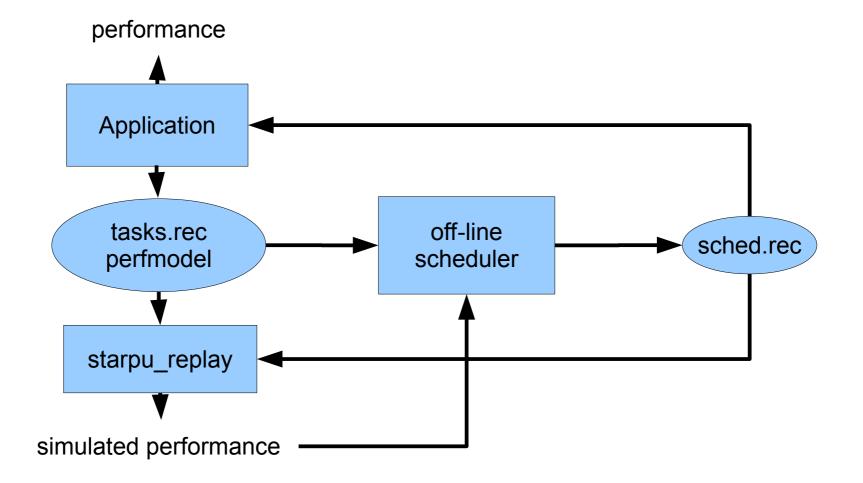
Beyond Simulation with SimGrid

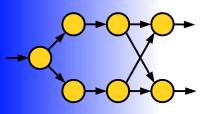
- Record application task graph
- Replay it with just starpu_replay





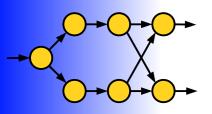
- Record application task graph
- Replay it, possibly with offline scheduling





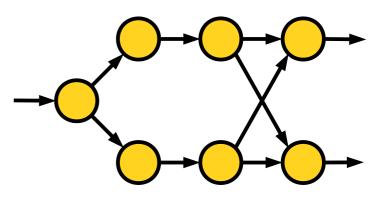
A task graph market?

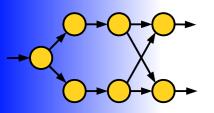
- Record application task graph
 - Varying application algorithms, dwarves, sizes
 - As many testcases for theoreticians
- Several levels of scheduling research
 - Just offline scheduling with tasks.rec
 - With high-level languages
 - Implement a real StarPU component
 - Run with starpu_replay
 - Care more about realworld conditions
 - Scheduler speed, still with starpu_replay
 - Run real applications in simulation
 - Run for real



Today's agenda

- Task graph scheduling
 - What is a runtime scheduler?
 - Beyond classical HEFT
 - Involving theoreticians
- Dividable tasks
- Distributed execution





How big should a task be?

Smaller task granularity

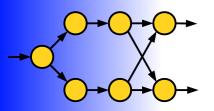
- Exposing parallelism
- Fine-grain load balancing

Large task granularity

• Needed by GPUs for efficiency

Making a compromise?

- Not very conclusive
- \rightarrow Adaptative task size?

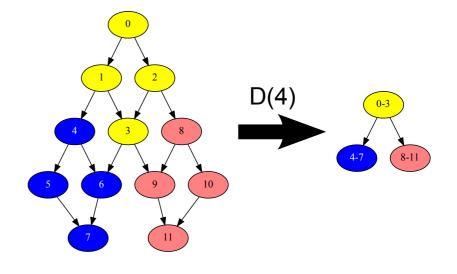


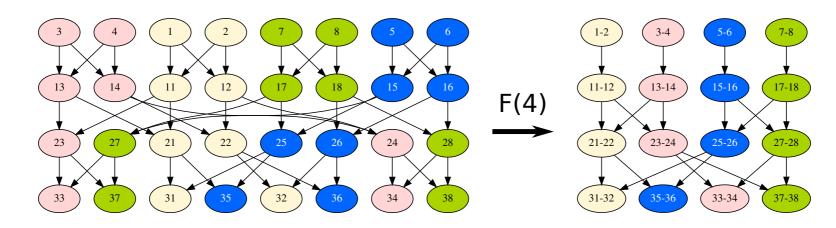
Gathering tasks

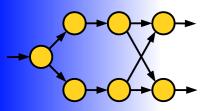
[RossignonPhD15]: Taggre

- Fine-grain task graph
- Gathering recipies
- Before submission to runtime

Allows to tinker with granularity without modifying application

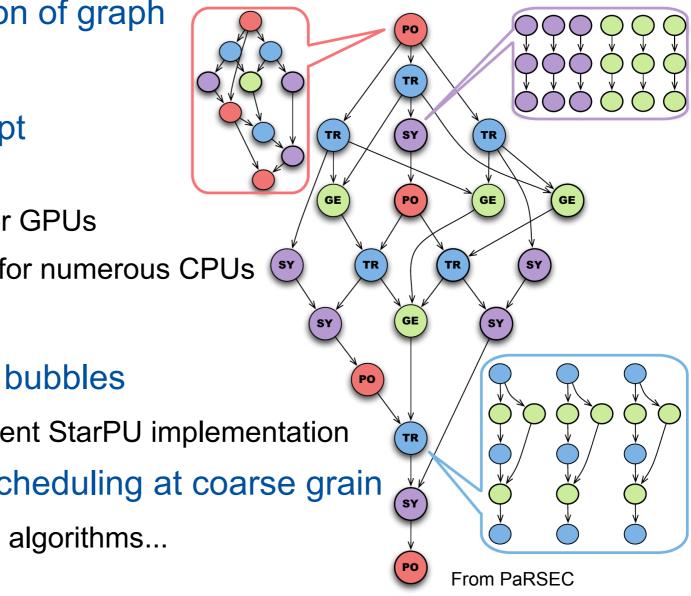


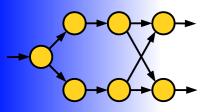




Recursive expression of graph

- **Dividable tasks**
- Dynamically adapt granularity
 - both big tasks for GPUs
 - and small tasks for numerous CPUs
- Similar to thread bubbles
 - Called so in current StarPU implementation
- Apply complex scheduling at coarse grain
 - For $O(n^2) / O(n^3)$ algorithms...

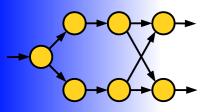


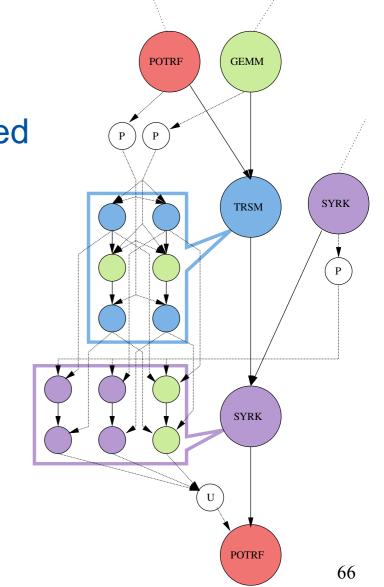


POTRF GEMM SYRK TRSM SYRK POTRF 65

Synchronization concerns

- Fork-join parallelism
- Hindered by synchronization induced by task graph



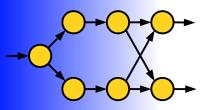


Synchronization concerns

- Fork-join parallelism
- Hindered by synchronization induced by task graph

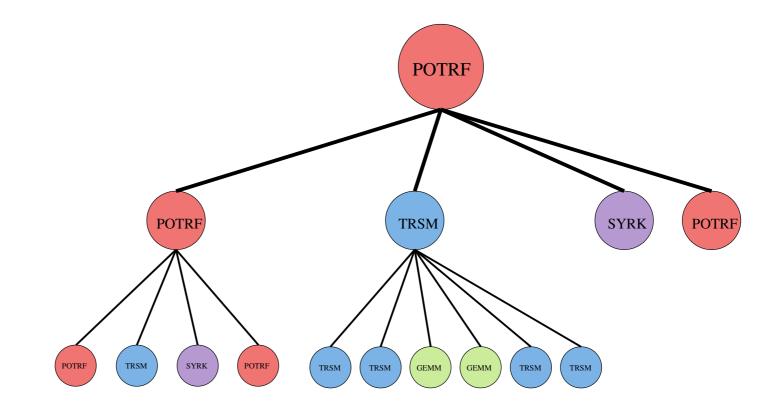
Multi-level data coherency

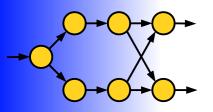
- Synchronization pseudo-tasks
 - Only when needed
- Result is exactly as appropriate



No extra synchronization

 \rightarrow Can consider task graph subdivision as a tree



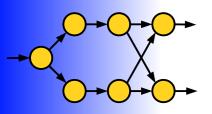


No extra synchronization

- \rightarrow Can consider task graph subdivision as a tree
- $\rightarrow\,$ Decide at will where and when to stop recursing

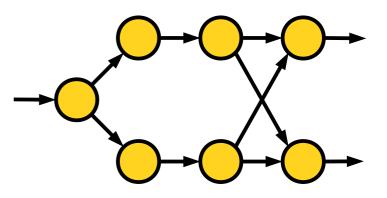
POTRF

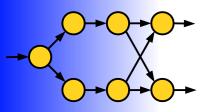
Actually Airbus' hmat approach How to nicely express it?



Today's agenda

- Task graph scheduling
 - What is a runtime scheduler?
 - Beyond classical HEFT
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- Dividable tasks
- Distributed execution

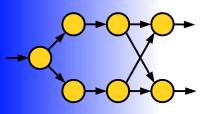




Distributed execution

Try to keep STF principle

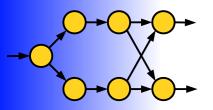
- Master-slave mode
 - Master unrolls whole task graph
 - Master schedules tasks between slaves
 - Just like scheduling between CPUs and GPUs
 - Limited scaling



Distributed execution

Try to keep STF principle

- Master-slave mode
- Completely distributed mode
 - Application provides data mapping
 - Task mapping according to data mapping
 - Task run on node which owns data written to
 - All nodes unroll the whole task graph
 - Only keep tasks they have to execute
 - Automatically generate communications



Distributed execution

Fully distributed with STF

```
For (k = 0 .. tiles - 1) {

POTRF(A[k,k])

for (m = k+1 .. tiles - 1)

TRSM(A[k,k], A[m,k])

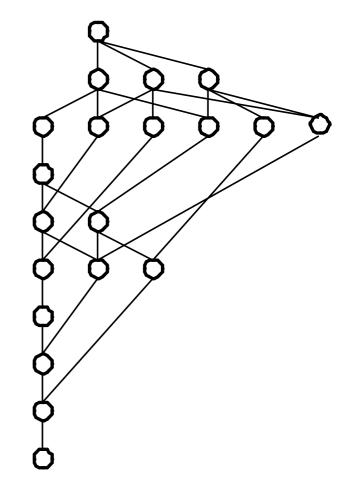
for (m = k+1 .. tiles - 1) {

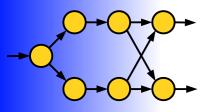
SYRK(A[m,k], A[m,m])

for (n = m+1 .. tiles - 1)

GEMM(A[m,k], A[n,k], A[n,m])

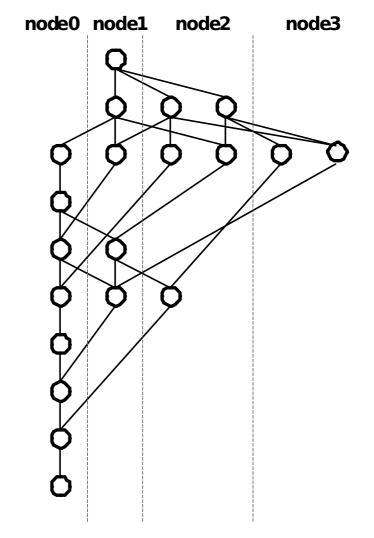
}
```

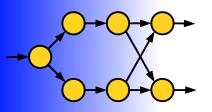




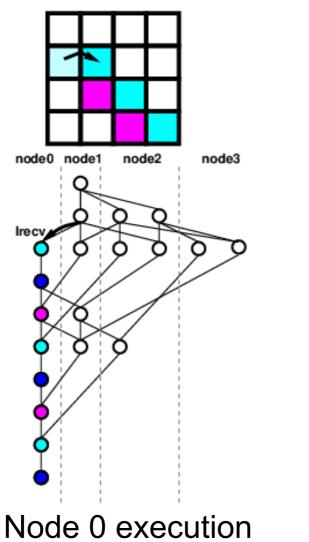
Fully distributed with STF

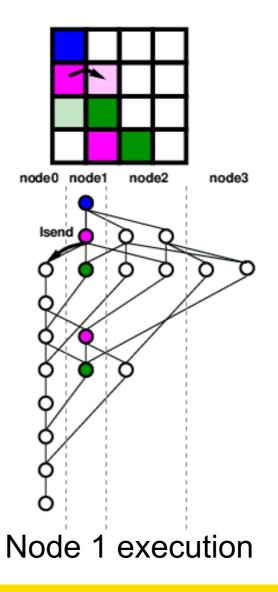
```
int get rank(int m, int n) { return ((m\%p)*q + n\%q); }
For (m = 0 ... tiles - 1)
  For (n = m \dots tiles - 1)
     set_rank(A[m,n], get_rank(m,n));
For (k = 0 ... tiles - 1) {
  POTRF(A[k,k])
  for (m = k+1 ... tiles - 1)
     TRSM(A[k,k], A[m,k])
  for (m = k+1 ... tiles - 1) {
     SYRK(A[m,k], A[m,m])
     for (n = m+1 \dots tiles - 1)
        GEMM(A[m,k], A[n,k], A[n,m])
```

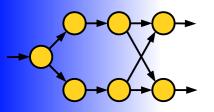




Fully distributed with STF

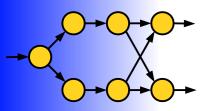




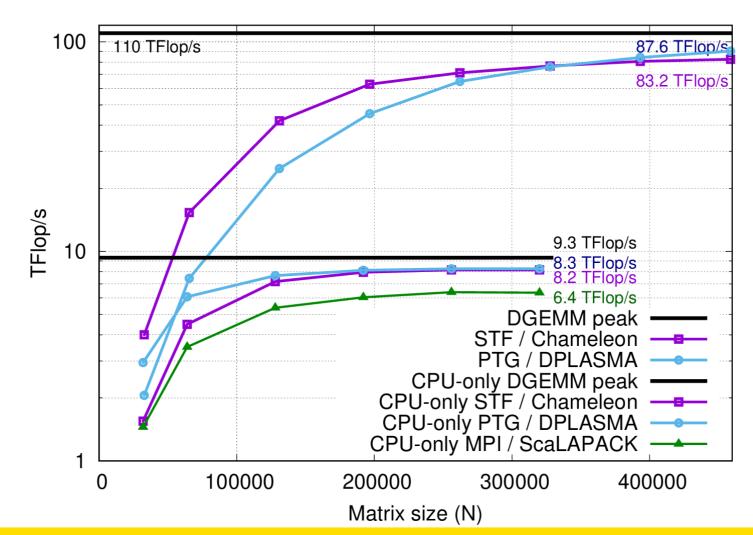


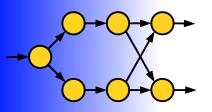
[SergentPhD16]

- Making it to scale
 - Caching values
 - Pruning task graph
- Require less memory
 - Throttling submission



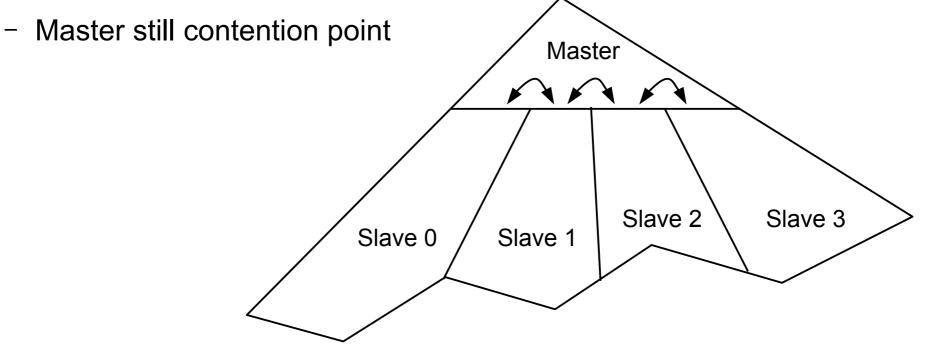
Result competitive with state of the art, over 144 nodes (1152 CPU cores, 288 GPUs)

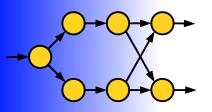




Could leverage dividable tasks

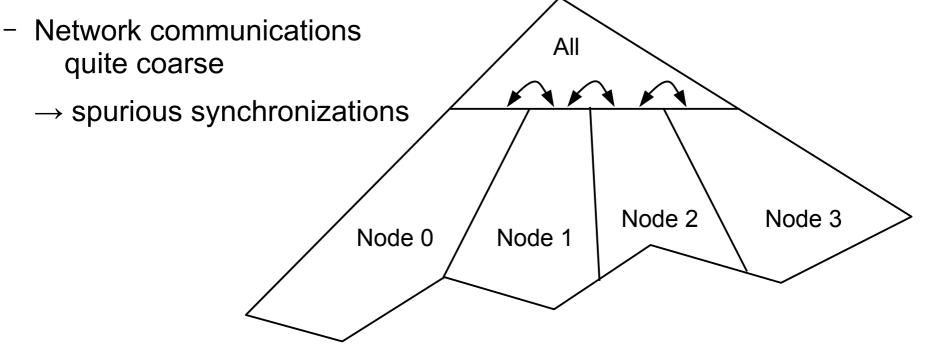
- ClusterSs from BSC
 - Master unrolls higher recursion levels, schedules result
 - Slaves unroll the rest

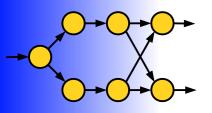




Could leverage dividable tasks

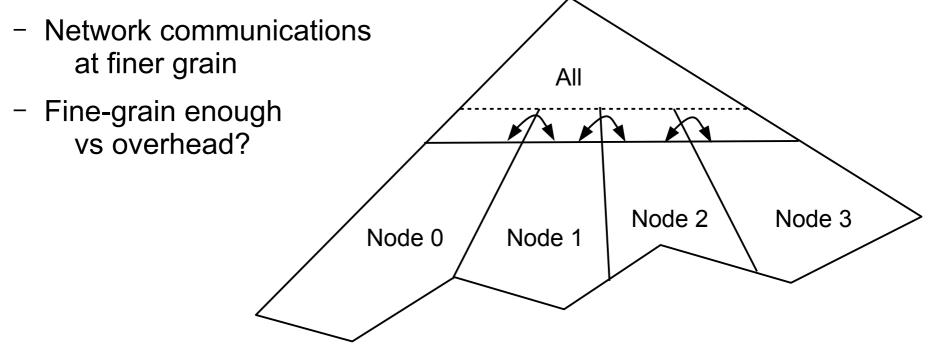
- DuctTeip from University of Uppsala
 - All nodes unroll higher recursion levels, determine task mapping
 - Nodes unroll their own remaining recursion

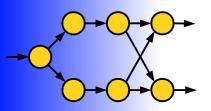




Could leverage dividable tasks

- Ideally?
 - All nodes unroll higher recursion levels, plus some margin
 - Nodes unroll their own remaining recursion





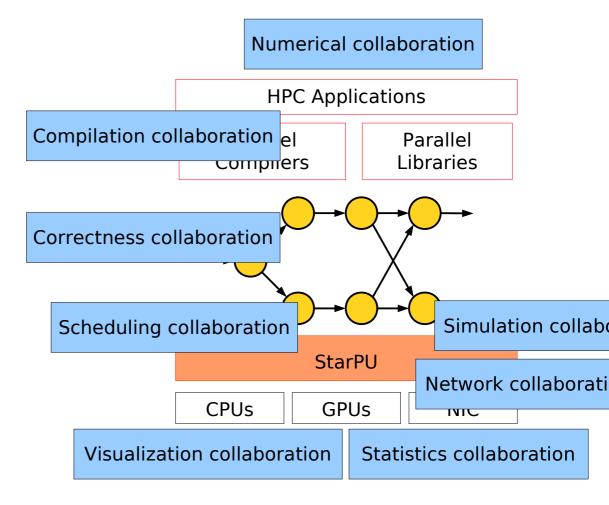
Conclusion

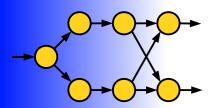
Heterogeneous applications on heterogeneous platforms

- Expressed with sequential-looking code (STF)
- Interesting scheduling problem
- Built bridge between scheduling theory and applications
- Dividable tasks to better address heterogeneity
- Fully distributed execution competitive

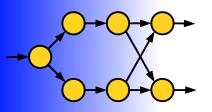
A lot of expertise gathered around StarPU

- Thanks to task graph
- Expressed with STF
- StarPU as a bridge between theory and applications



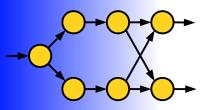


Conclusion



Perspectives

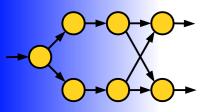
- Programming interface (OpenMP?, dividable tasks, BigData)
- Scheduling
 - Tasks
 - Memory ([ArrasPhD15])
 - Moldable tasks
- Composition (runtimes, programming paradigms)
- Leveraging compilers
- Visualization
- Model checking
- Fault tolerance of distributed execution
 - Checkpoint / restart + distributed load balancing [LionPhD]



Thanks!

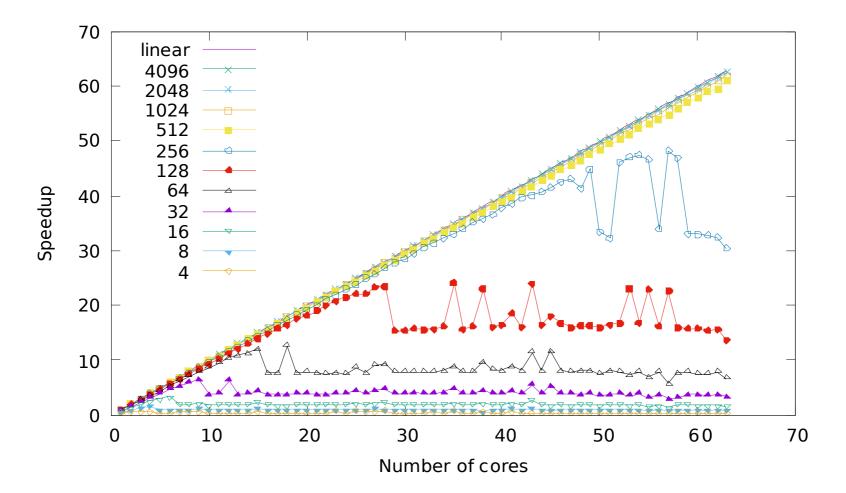
[AugonnetPhD11] [ArrasPhD15] [RossignonPhD15] [SergentPhD16] [KumarPhD17] [LionPhD]

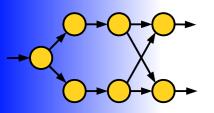
- ANR ProHMPT'09, MEDIAGPU'10, FP3C'10, SONGS'12, SOLHAR'13
- IPL HAC-SPECIS'16, HPC-BigData'18
- EU PEPPHER'10, EXA2PRO'18
- Rapid Hi-BOX'13
- MORSE associate-team
- Used by AL4SAN, Chameleon, ExaGeoStat, FLUSEPA, HiCMA, hmat, KSVD, MAGMA, MaPHyS, MOAO, PaStiX, QDWH, qr_mumps, ScalFMM, SCHNAPS, SignalPU, SkePU, STARS-H



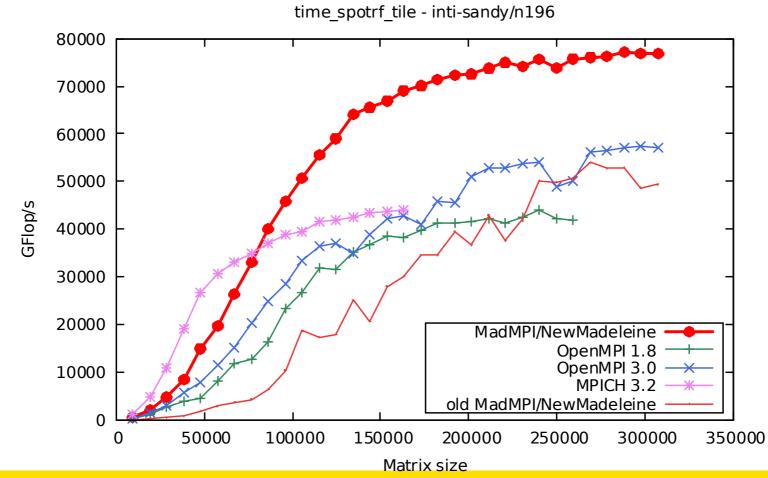
Lower bound due to runtime overhead

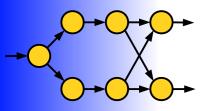
• Proposed by Martin Tillenius





- Classical MPI implementations not used to irregular communications
- Collaborating with A. Denis

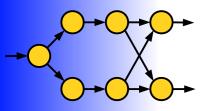




All nodes unrolling all loops expensive

• Task graph pruning

```
#define GEMM(A, B, C) \
if (get rank(A) == self || get rank(B) == self || get rank(C) == self) \
GEMM(A, B, C);
 For (k = 0 ... tiles - 1) {
    POTRF(A[k,k])
   for (m = k+1 \dots tiles - 1)
      TRSM(A[k,k], A[m,k])
   for (m = k+1 ... tiles - 1) {
      SYRK(A[m,k], A[m,m])
      for (n = m+1 \dots tiles - 1)
         GEMM(A[m,k], A[n,k], A[n,m])
```



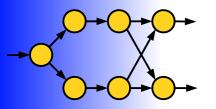
All nodes unrolling all loops expensive

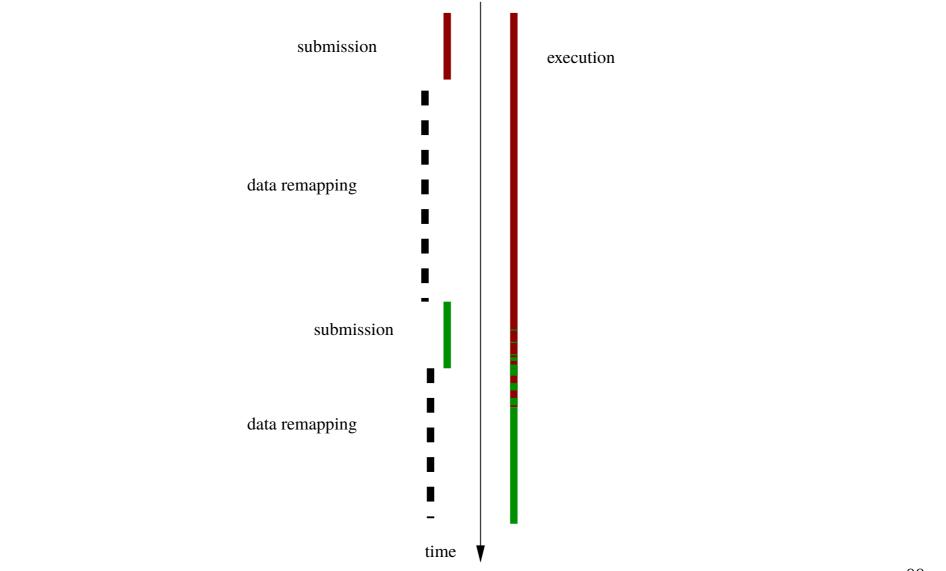
• Could use source-to-source compilation to rework loop nest

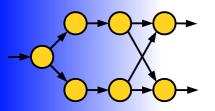
```
- Polyhedral analysis, to get
For (k = 0 .. tiles - 1) {
    POTRF(A[k,k])
    for (m = k+1 .. tiles - 1)
        TRSM(A[k,k], A[m,k])
    for (m = k+1 .. tiles - 1) {
        SYRK(A[m,k], A[m,m])
        for (n = m+1+xxx .. tiles - 1 step Q)
            GEMM(A[m,k], A[n,k], A[n,m])
    }
```

Reduce complexity from O(n³) to O(n³/Q)

Dynamic data remapping







Checkpoint / restart

